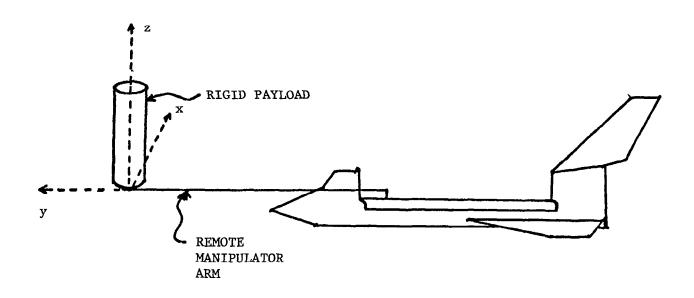
IDENTIFICATION AND CONTROL OF SPACECRAFT

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IDENTIFICATION & CONTROL OF SPACECRAFT

- * THE PROBLEM
- * CONTROL
 - CLASSICAL
 - MODERN
- * IDENTIFICATION

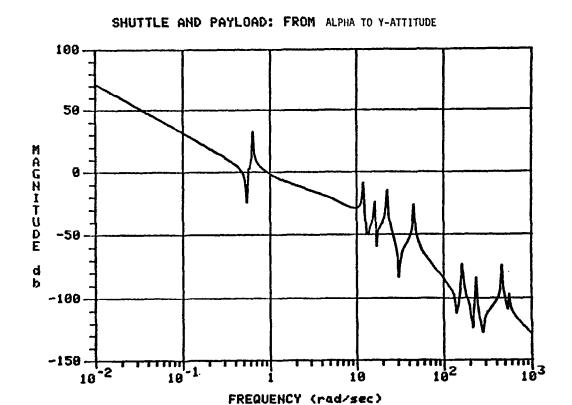


CLASSICAL CONTROL

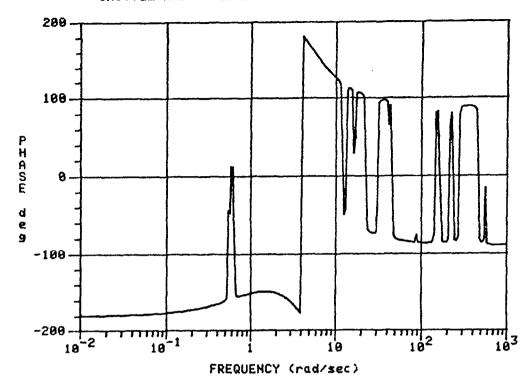
- * SCALAR FEEDBACK DESIGN
- * LEAD-LAG CONTROLLER

$$K(s) = \frac{k(s+a)}{(s^2+2\zeta b+b^2)}$$

* BANDWIDTH AROUND 1 RAD/SEC

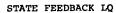


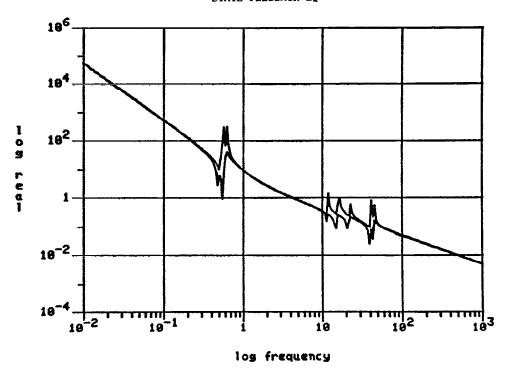


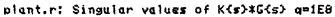


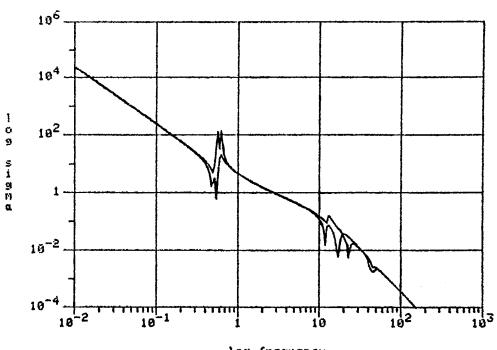
MODERN CONTROL

- * START WITH FULL-STATE DESIGN
 - GOAL: MINIMIZE PAYLOAD ATTITUDE ERRORS
 - ITERATE ON CONTROL PENALTY TO ACHIEVE BW OF 5 R/s
- * DESIGN FILTER TO RECOVER LQ RESPONSE
 - USE STEIN/DOYLE ROBUSTNESS RECOVERY RESULTS
- * TEST ROBUSTNESS



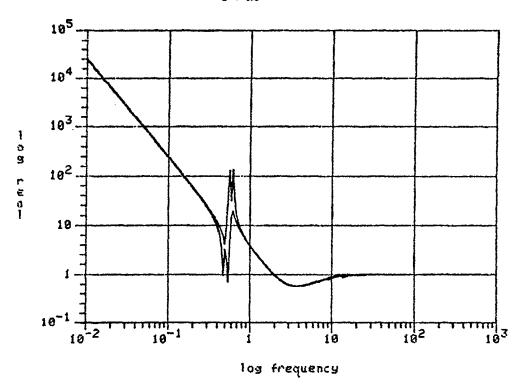


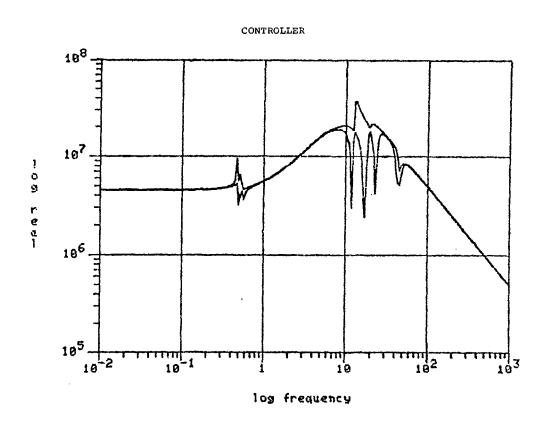




log frequency

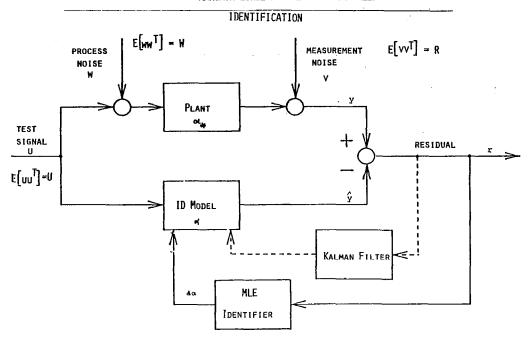






IDENTIFICATION

MAXIMUM LIKELIHOOD ESTIMATION (MLE)



MODEL STRUCTURE

- STATE SPACE
 - n. Modes (2x2 blocks)
 - m INPUTS
 - P OUTPUTS

$$\begin{bmatrix} \vdots \\ \overline{\mathbf{x}}_{\mathbf{i}} \\ \overline{\mathbf{x}}_{\mathbf{i}} \\ \vdots \\ 0 \end{bmatrix} (t) = \begin{bmatrix} \vdots \\ 0 & 1 & 1 \\ \vdots & 0 & 1 \\ 0 & 1 & 1 \\ \vdots & 0 & 1 \end{bmatrix} \begin{bmatrix} \vdots \\ \mathbf{x}_{\mathbf{i}} \\ \mathbf{x}_{\mathbf{i}} \\ \vdots \\ \vdots \end{bmatrix} (t) + \begin{bmatrix} \vdots \\ 0 & 1 \\ \vdots \\ 0 & 1 \end{bmatrix} [\underline{u}(t) + \mathbf{w}(t)]$$

$$\begin{bmatrix} y \end{bmatrix} (e) = \begin{bmatrix} \dots & c_i & 0 & \dots \end{bmatrix} \begin{bmatrix} \vdots \\ \overline{X_i} \\ \underline{X_i} \\ \vdots \end{bmatrix} (e) + \begin{bmatrix} v \end{bmatrix} (e)$$

• FREQUENCY DOMAIN

$$y(s) = (G(s)(u(s)+w(s)) + v(s)$$

$$G_{\star}(s) = \sum_{i=1}^{n_{\star}} \frac{c_{i} b_{i}^{\mathsf{T}}}{s^{2} + 2 \epsilon_{i} \omega_{i} s + \omega_{i}^{2}}$$

PARAMETER VECTOR

$$\alpha_{\star} = \{\omega_{\star_{1}}^{2}, \ 2\zeta_{\star_{1}}\omega_{\star_{1}}, \ b_{\star_{1}}, \ c_{\star_{1}}; \ i = 1, \ \ldots, \ n_{\star}\}$$

MLE IDENTIFICATION SETUP

• RESIDUAL DEFINITION

$$r_k \stackrel{\triangle}{=} y(kT) - \hat{y}(kT)$$

• LIKELIHOOD FUNCTION (NEGATIVE LOG)

$$\begin{aligned} \mathbf{L}(\alpha) & \stackrel{\triangle}{=} & \sum_{\mathbf{k}=0}^{N} & \mathbf{L}_{\mathbf{k}}(\alpha) \\ & \stackrel{\triangle}{=} & \sum_{\mathbf{k}=0}^{N} & \frac{1}{2} \left[\text{log det } \mathbf{S}_{\mathbf{k}} + \mathbf{r}_{\mathbf{k}}^{\mathbf{T}} \, \mathbf{S}_{\mathbf{k}} \, \mathbf{r}_{\mathbf{k}} \right] \end{aligned}$$

WHERE

$$\alpha \triangleq \{\omega_{i}^{2}, \ 2c_{i}\omega_{i}, \ b_{i}, \ c_{i}; \ i=1, \ \ldots, \ n\} = \text{UNKNOWN PARAMETERS}$$

$$s_{k} \triangleq E_{\alpha}\{r_{k}r_{k}^{T}\} = \text{PREDICTED RESIDUAL COVARIANCE}$$

MLE IDENTIFICATION SOLUTION

PARAMETER ESTIMATE (THEORETICAL)

$$\hat{\alpha} \stackrel{\Delta}{=} ARG \left\{ \begin{array}{l} MIN \\ \alpha \end{array} L(\alpha) \right\} = PARAMETER ESTIMATE$$

• ITERATIVE ALGORITHMS

GRADIENT:
$$\hat{\alpha}_{J+1} = \hat{\alpha}_J - \epsilon_J \nabla L(\hat{\alpha}_J)$$

Newton-Rhapson: $\hat{\alpha}_{J+1} = \hat{\alpha}_J - \left[\nabla^2 L(\hat{\alpha}_J)\right]^{-1} \nabla L(\hat{\alpha}_J)$

WHERE

$$\nabla L(\alpha) \triangleq \frac{9L}{9\alpha} (\alpha)$$

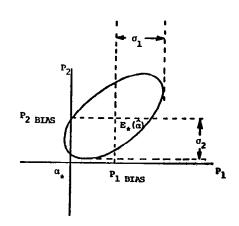
$$\nabla^2 L(\alpha) \triangleq \frac{32L}{3\alpha^2} (\alpha)$$

IDENTIFICATION ACCURACY ISSUES

- SYSTEMATIC ERRORS: E_{*}(α) α_{*}
 - MODEL-ORDER MISMATCH
 - TEST SIGNAL AMPLITUDE AND SHAPING
 - SYSTEMATIC DISTURBANCES
 - -SENSOR/ACTUATOR MODEL ERRORS



- RANDOM DISTURBANCES AND SENSOR NOISE
- TEST SIGNAL AMPLITUDE AND SHAPING
- IDENTIFICATION TIME INTERVAL



STEADY-STATE IDENTIFIABILITY ANALYSIS (YARED)

EXPECTED LIKELIHOOD FUNCTION

$$I^*(\alpha) = E_* \left\{ L_K^{(\alpha)} \right\}$$

$$= \frac{1}{2} \left[LOG DET S + TR(S^{-1}S_*) \right]$$

Residual Covariances

$$S = E_{\alpha} \left\{ r_{K} r_{K}^{T} \right\} = KALMAN FILTER PREDICTED RESIDUAL COVARIANCE MATRIX$$

$$S_* = E_* \left\{ r_K r_K^T \right\} = Actual residual covariance matrix$$

Note: S. and $1^{\bullet}(\alpha)$ can only be computed when the true plant parameters are known.

EXPECTED MLE IDENTIFICATION SOLUTION

• EXPECTED PARAMETER ESTIMATE (THEORETICAL)

$$\hat{\alpha} = \frac{\Delta}{\pi} E * \left\{ \alpha \right\} = ARG \left\{ \begin{pmatrix} MIN \\ \alpha \end{pmatrix} I^{\bullet}(\alpha) \right\}$$

• ITERATIVE ALGORITHMS

GRADIENT:
$$\hat{\alpha}_{\bullet,j+1} = \hat{\alpha}_{\bullet,j} - \epsilon_{j} \nabla I^{\bullet}(\hat{\alpha}_{\bullet,j})$$

Newton-Rhapson: $\hat{\alpha}_{\bullet,j+1} = \hat{\alpha}_{\bullet,j} - \left[\nabla^{2}I^{\bullet}(\hat{\alpha}_{\bullet,j})\right]^{-1} \nabla I^{\bullet}(\hat{\alpha}_{\bullet,j})$

WHERE
$$\nabla I^*(\alpha) = \frac{\partial^2 I^*}{\partial \alpha}(\alpha)$$

$$\nabla^2 I^*(\alpha) = \frac{\partial^2 I^*}{\partial \alpha^2}(\alpha)$$

STEADY STATE IDENTIFICATION ACCURACY

- SYSTEMATIC ERRORS (BIASES)
 - PARAMETER ERRORS

$$\alpha_{\text{BIAS}} \stackrel{\triangle}{=} \hat{\alpha}_{\star} - \alpha_{\star} = 0$$
 when no model mismatch

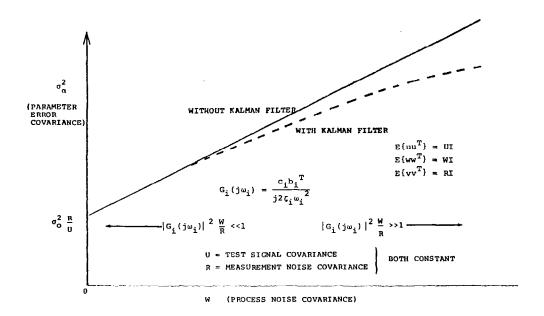
- INFORMATION MEASURE (YARED)

$$I(\alpha_{\star}; \hat{\alpha}_{\star}) \stackrel{\Delta}{=} I^{\star}(\hat{\alpha}_{\star}) - I^{\star}(\alpha_{\star}) \geq 0$$

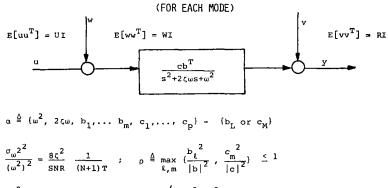
STOCHASTIC ERRORS

$$\begin{split} \mathsf{C}_{\mathsf{OV}} \{ \hat{\mathbf{a}} \} & \stackrel{\triangle}{=} & \mathsf{Lim}_{\mathsf{N}} \quad \mathsf{E}_{\bullet} \quad \left\{ (\hat{\alpha} - \hat{\alpha}_{\bullet}) \cdot (\hat{\alpha} - \hat{\alpha}_{\bullet})^{\mathsf{T}} \right\} \\ & \stackrel{\triangle}{=} & \left[\nabla^2 \mathsf{I}^{\bullet} \cdot (\hat{\alpha}_{\bullet}) \right]^{-1} \quad \mathsf{E}_{\bullet} \left\{ \left[\frac{\partial \mathsf{I}}{\partial \alpha} \cdot (\hat{\alpha}_{\bullet}) \right] \cdot \left[\frac{\partial \mathsf{I}}{\partial \alpha} \cdot (\hat{\alpha}_{\bullet}) \right] \cdot \left[\nabla^2 \mathsf{I}^{\bullet} \cdot (\hat{\alpha}_{\bullet}) \right]^{-1} \cdot \prod_{(\mathsf{N}+1)} 2 \right. \\ & = & \left[\nabla^2 \mathsf{I}^{\bullet} \cdot (\hat{\alpha}_{\bullet}) \right]^{-1} \cdot \frac{1}{(\mathsf{N}+1)} \quad \text{when no model mismatch} \end{split}$$

STOCHASTIC ERROR WITH PROCESS NOISE



SIMPLIFIED IDENTIFICATION ACCURACY ANALYSIS



$$\frac{\sigma_{2\zeta\omega}^{2}}{(2\zeta\omega)^{2}} = \frac{4}{\text{SNR}} \frac{1}{(\text{N+1}) \text{ T}} ; \quad \text{SNR} \triangleq \begin{cases} \frac{|c|^{2}|b|^{2}}{4\zeta\omega^{3}} & \frac{U}{R} \\ \zeta\omega & \frac{U}{W} \end{cases}$$
 (MEAS. NOISE)

$$\frac{\sigma_{b\ell}^{2}}{b_{\ell}^{2}} = \frac{1}{SNR} \left[\frac{|b|^{2}}{b_{\ell}^{2}} + \rho^{-1} \right] = \frac{1}{(N+1)T}$$

$$\frac{\sigma_{cm}^{2}}{\sigma_{m}^{2}} = \frac{1}{SNR} \left[\frac{|c|^{2}}{\sigma_{m}^{2}} + \rho^{-1} \right] \frac{1}{(N+1)T}$$

NOTE: THIS ANALYSIS ASSUMES THAT $\omega T << 1$, $\varsigma << 1$ AND APPLIES FOR EACH MODE

NUMERICAL RESULTS

PROBLEM SIZE

12 MODES (=22 - 8R.B. - 2 SMALL)
2 INPUTS (α, β GIMBAL ANGLES)
2 OUTPUTS (x, γ ATTITUDES)
60 PARAMETERS (=12 MODES x 5 PARAMETERS/MODE)

• TEST SIGNAL, NOISE STATISTICS

SAMPLE TIME: T = 0.1 SECTEST SIGNAL: $U = 4000(\text{IN-LB})^2$ PROCESS NOISE: $W = 40(\text{IN-LB})^2$ MEASUREMENT NOISE: $R = 4 \times 10^{-12} \text{ RAD}^2$

• WORST-CASE RELATIVE ERRORS AT TIME (MODE 9):

PARAMETER	1_SEC	14 SEC	0.39 HRS
ω ²	0.0265	0.007	0.0007
2 ζω	3.75	1.0	0.1
ь ₁	13.6	3.6	0.36
c ₁	2.7	0.7	0.07
c ₂	11.7	3.1	0.31

SUMMARY

- CONTROL PROBLEM
 - MODERN LQ CONTROL DESIGN WITH ROBUSTNESS RECOVERY PRODUCES ROBUST CONTROLLERS FOR LSS
 - ACCURATE ID ALLOWS A FIVE-FOLD INCREASE IN LOOP BW
- IDENTIFICATION PROBLEM
 - STRUCTURAL MODES MAY BE IDENTIFIED ONE AT A TIME FOR SMALL DAMPING
 - LSS ID W/O KF
 - -- GREATLY REDUCES PARAMETER BIASES
 - -- GIVES ONLY MODEST INCREASE IN STOCHASTIC ERRORS
 - RELATIVE ERRORS IN PARAMETERS AFTER ID ARE SMALLER FOR FREQUENCY THAN FOR DAMPING OR MODE SHAPES
- OPEN ISSUE: HOW ACCURATE MUST ID BE FOR ROBUST CONTROL DESIGN?